

# Supercriticality for Annealed Approximations of Boolean Networks

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## Abstract

We consider a model proposed by Derrida and Pomeau (1986) and recently studied by Chatterjee and Durrett (2009); it is defined as an approximation to S. Kauffman's boolean networks (1969). The model starts with the choice of a random directed graph on  $n$  vertices; each node has  $r$  input nodes pointing at it. A discrete time threshold contact process is then considered on this graph: at each instant, each site has probability  $q$  of choosing to receive input; if it does, and if at least one of its inputs were occupied by a 1 at the previous instant, then it is labeled with a 1; in all other cases, it is labeled with a 0.  $r$  and  $q$  are kept fixed and  $n$  is taken to infinity. Improving a result of Chatterjee and Durrett, we show that if  $qr > 1$ , then the time of persistence of the dynamics is exponential in  $n$ .

*Keywords:* boolean networks, threshold contact process

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# 1 Introduction

In this paper we consider a threshold contact process on a random graph. Let  $r \in \mathbb{N}$  and  $q \in [0, 1]$ . For  $n \in \mathbb{N}$ ,  $n > r$ , let  $V_n = \{1, \dots, n\}$ . For each  $x \in V_n$ , choose  $r$  distinct points  $y_1(x), \dots, y_r(x)$  in  $V_n - \{x\}$ ; this choice is made uniformly among all  $\frac{(n-1)!}{(n-r-1)!}$  possibilities and independently for each  $x \in V_n$ . Let  $E_n = \{(y_i(x), x) : x \in V_n, 1 \leq i \leq r\}$  and call  $G_n = (V_n, E_n)$  the graph thus obtained, a random directed graph on  $n$  vertices and in-degree equal to  $r$ .

Once  $G_n$  is chosen, it remains fixed and we consider a discrete time Markov chain with state space  $\{0, 1\}^{V_n}$  and initial configuration  $\xi_0 \in \{0, 1\}^{V_n}$ , which will be deterministic for all our purposes. Let  $\{B_t^x : x \in V_n, t \geq 1\}$  be a family of independent Bernoulli random variables with parameter  $q$ ; given  $\xi_t \in \{0, 1\}^{V_n}$ , we put

$$\xi_{t+1}(x) = \begin{cases} 1 & \text{if } B_{t+1}^x = 1 \text{ and } \sum_{i=1}^r \xi_t(y_i(x)) > 0; \\ 0 & \text{otherwise.} \end{cases}$$

When  $B_t^x = 1$ , we say that  $x$  *receives input* at time  $t$ ; therefore, a vertex is set to 1 if and only if it receives input at that time and at least one of its input vertices  $y_1(x), \dots, y_r(x)$  was set to 1 at the previous time. Given  $A \subset V_n$ ,  $\xi_t^A$  denotes the chain with initial configuration  $\xi_0^A = \mathbf{1}_A$ , the indicator function of  $A$ . We write  $\xi_t^x$  instead of  $\xi_t^{\{x\}}$ .  $\mathbb{P}_n$  will denote the probability measure both for the choice of  $G_n$  and for the family  $\{B_t^x\}$  (they are of course taken independently).

This setting was proposed by Derrida and Pomeau ([2]) as an “annealed approximation” to the less tractable S. Kauffman’s boolean networks ([4]). These are cellular automata with rules of evolution randomly chosen, intended to model the interactions of genes in a cell. We refer the reader to [1] for the detailed scientific background, including the relationship between the threshold contact process defined above and Kauffman’s original model.

For fixed  $n$  and any initial configuration  $\xi_0 \in \{0, 1\}^{V_n}$ , with probability one the threshold contact process eventually reaches the absorbing configuration in which all vertices are in state 0. The main object of investigation both in [2] and [1] is the distribution of this random time as a function of  $n$ , in particular as  $n \rightarrow \infty$ . Define  $\rho = \rho(q, r)$  as the survival probability for a branching process in which individuals have probability  $q$  of having  $r$  children and  $1 - q$  of having none. In [1] the following is proved.

**Theorem 1.1** [Chatterjee and Durrett 2009]

(i.) If  $qr > 1$ , then for every  $\eta > 0$  there exist  $c > 0, b \in (0, 1)$  such that, as  $n \rightarrow \infty$ ,

$$\inf_{0 \leq t \leq e^{cn^b}} \mathbb{P}_n \left( \frac{|\xi_t^{V_n}|}{n} \geq \rho - \eta \right) \rightarrow 1.$$

(ii.) If  $q(r-1) > 1$ , then for every  $\eta > 0$  there exists  $c > 0$  such that, as  $n \rightarrow \infty$ ,

$$\inf_{0 \leq t \leq e^{cn}} \mathbb{P}_n \left( \frac{|\xi_t^{V_n}|}{n} \geq \rho - \eta \right) \rightarrow 1.$$

In this paper we improve this result.

**Theorem 1.2** *If  $qr > 1$ , then for every  $\eta > 0$  there exists  $c > 0$  such that, as  $n \rightarrow \infty$ ,*

$$\inf_{0 \leq t \leq c^n} \mathbb{P}_n \left( \frac{|\xi_t^{V_n}|}{n} \geq \rho - \eta \right) \rightarrow 1.$$

To explain why this result is to be expected and, in particular, the link with the mentioned branching process, we introduce the time dual of the model. Fix a realization of  $G_n = (V_n, E_n)$  and  $\{B_t^x : x \in V_n, t \geq 1\}$ , define  $\hat{E}_n$  as the set of directed edges obtained by inverting the edges of  $E_n$  and  $\hat{G}_n = (V_n, \hat{E}_n)$ . Fix  $T > 0$  and put  $\hat{B}_t^{x,T} = B_{T-t}^x$  for  $0 \leq t < T$ . Given  $A \subset V_n$ , define  $\hat{\xi}_0^{A,T} = \mathbf{1}_A$  and, for  $0 \leq t < T$ ,

$$\hat{\xi}_{t+1}^{A,T}(x) = \begin{cases} 1 & \text{if for some } z, i, \text{ we have } y_i(z) = x, \hat{\xi}_t^{A,T}(z) = 1 \text{ and } \hat{B}_t^{z,T} = 1; \\ 0 & \text{otherwise.} \end{cases}$$

When  $\hat{\xi}_t^T(z) = 1$  and  $\hat{B}_t^{z,T} = 1$ , we say that  $z$  gives birth at time  $t$ , in which case  $y_1(z), \dots, y_r(z)$  will all be in state 1 at  $t + 1$ . We have the *duality equation*

$$\{\xi_T^A \cap B \neq \emptyset\} = \{\hat{\xi}_T^{B,T} \cap A \neq \emptyset\}$$

(we abuse notation associating  $\xi \in \{0, 1\}^{V_n}$  with  $\{x \in V_n : \xi(x) = 1\}$ ). Since we will only work with the dual process, we will drop the superscript  $T$  and assume that  $\hat{\xi}_t^A$  is defined for all positive times with the evolution rule defined above.

Now, assume that  $n$  is very large with respect to  $r$ . If  $g$  is another integer that is much larger than  $r$  and much smaller than  $n$ , then with high probability the set

$$\{z \in V_n : \text{for some } k \leq g \text{ and } z_1, \dots, z_k \in V_n, \text{ we have } x \rightarrow z_1 \rightarrow \dots \rightarrow z_k \rightarrow z \text{ in } \hat{G}_n\}$$

will simply be a directed tree of degree  $r$  rooted in  $x$ . Conditioning on this event, the evolution of  $|\hat{\xi}_t^x|$  up to time  $g$  will be exactly that of the branching process mentioned before Theorem 1.1. In addition, it is not difficult to see that, without any conditioning,  $|\hat{\xi}_t^x|$  is stochastically dominated by such a process. This remarks clarify why the model exhibits two phases in exact correspondence with the branching process. If the expected offspring size  $qr \leq 1$ , then  $\hat{\xi}_t^x$  dies out faster than the corresponding subcritical branching process, and the primal  $\xi_t^{V_n}$  rapidly reaches the zero state. On the other hand, if  $qr > 1$ , the above theorem states that the system survives for a time that is exponentially large in  $n$ , characterizing the supercritical regime.

In the treatment of the dynamics, our proof is basically an exact repetition of that of [1]. What we do different is a more careful examination of the random graph. In order to argue that the confinement to a finite graph takes a long time to affect the dynamics, Chatterjee and Durrett prove an isoperimetric inequality that states that, if  $m$  is small in relation to  $n$ , then with high probability there are no subsets  $A \subset V_n$  of size  $m$  such that the ‘‘influence set’’  $\{y_i(x) : 1 \leq i \leq r, x \in V_n\}$  has much less than  $rm$  elements. We push this argument further and control the influence set along several generations rather than only the first one.

## 2 Proof of Theorem 1.2

In all the results and proofs in this section, we assume that  $qr > 1$ . Also, once and for all we fix  $\tilde{q} < q$  such that  $\tilde{q}r > 1$ ,  $\delta < (\tilde{q}r - 1) \wedge 1$  and  $g \in \mathbb{N}$  such that  $(\tilde{q}r - 1 - \delta)(\tilde{q}r)^{g-1} > 1 + \delta$ .

The following lemma is proved in [1]; see Lemma 2.2 and Equation (2.14) in that paper.

**Lemma 2.1** For every  $\eta > 0$  there exist  $a > 0$ ,  $b \in (0, 1)$  such that, as  $n \rightarrow \infty$ ,

$$\mathbb{P}_n \left( \frac{|\{x : |\hat{\xi}_{\lfloor a \log n \rfloor}^x| > n^b\}|}{n} > \rho - \eta \right) \rightarrow 1.$$

We now introduce some definitions and notation. Given  $m \in \mathbb{N}$ , let

$$T_m^i = \{1, \dots, m\} \times \{1, \dots, r\}^i, \quad 0 \leq i \leq g,$$

$$T_m = \cup_{i=0}^g T_m^i.$$

For  $\sigma = (\sigma_0, \dots, \sigma_i)$ ,  $\sigma' = (\sigma'_0, \sigma'_1, \dots, \sigma'_j) \in T_m$ , we say  $\sigma \prec \sigma'$  either if  $i < j$  or if  $i = j$  and  $\sigma$  is less than  $\sigma'$  in lexicographic order. With this order, we can take an increasing enumeration  $T_m = \{\sigma^1, \dots, \sigma^{(1+r+\dots+r^g)m}\}$ . Then,  $T_m^0 = \{\sigma^1, \dots, \sigma^m\}$  and, for  $i \geq 1$ ,  $T_m^i = \{\sigma^{(1+r+\dots+r^{i-1})m+1}, \dots, \sigma^{(1+r+\dots+r^i)m}\}$ .

Next, we endow  $T_m$  with directed edges by setting

$$\sigma \rightarrow \sigma' \text{ if and only if } \sigma = (\sigma_0, \dots, \sigma_i), \sigma' = (\sigma_0, \dots, \sigma_i, \sigma'_{i+1}) \text{ for some } i.$$

$T_m$  is thus the disjoint union of  $m$  rooted, directed and  $r$ -regular trees, each with  $g$  generations above the root.  $\{0, 1\}^{T_m}$  will be called the space of configurations. Given vertex  $\sigma \in T_m$  and configuration  $\psi \in \{0, 1\}^{T_m}$ ,  $\psi(\sigma) \in \{0, 1\}$  will denote the value of  $\psi$  at  $\sigma$ .

Let  $A \subset V_n$  with  $|A| = m$ . We can enumerate  $A = \{x_1, \dots, x_m\}$  in the order of the indices of  $V_n$ . Given  $\sigma = (\sigma_0, \dots, \sigma_i) \in T_m$  with  $i > 0$ , let  $z^\sigma = y_{\sigma_i}(y_{\sigma_{i-1}}(\dots(y_{\sigma_1}(x_{\sigma_0}))\dots))$ . Finally, define

$$\mathcal{A}^\sigma = \{z^{\sigma'} \in T_m : \sigma' \prec \sigma\}.$$

We now present an algorithm to construct a configuration  $\psi = \psi(A) \in \{0, 1\}^{T_m}$  from  $A$ .

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for  $j = 1$  to  $m$  set  $\psi(\sigma^j) = 0$ ;
for  $j = m + 1$  to  $(1 + r + \dots + r^g)m$ 
  if  $[\psi((\sigma^j)_0, (\sigma^j)_1, \dots, (\sigma^j)_l) = 1 \text{ for some } l]$  or  $[z^{\sigma^j} \notin \mathcal{A}^{\sigma^j}]$ 
    then set  $\psi(\sigma^j) = 0$ 
  else set  $\psi(\sigma^j) = 1$ 

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In words, vertices are inspected in order; the roots are all set to 0 and the other vertices are set to 0 either if one of their ancestors has already been marked with a 1 or if their image under the map  $\sigma \mapsto z^\sigma$  has never been seen before; otherwise they are set to 1.

**Lemma 2.2** Given  $A \subset V_n$  with  $|A| = m$  and  $\sigma^{i_1}, \dots, \sigma^{i_k} \in T_m$ ,

$$\mathbb{P}_n([ \psi(A) ](\sigma^{i_1}) = \dots = [ \psi(A) ](\sigma^{i_k}) = 1) \leq \left( \frac{m + rm + \dots + r^g m}{n - r} \right)^k.$$

*Proof.* There is no loss of generality in assuming that  $\sigma^{i_a} \prec \sigma^{i_b}$  when  $a < b$ . We then have

$$\mathbb{P}_n([ \psi(A) ](\sigma^{i_k}) = 1 \mid [ \psi(A) ](\sigma^{i_1}) = \dots = [ \psi(A) ](\sigma^{i_{k-1}}) = 1) \leq \frac{m + rm + \dots + r^g m}{n - r}.$$

Indeed, let  $\Theta^{i_k}$  denote the event that none of the ancestors of  $\sigma^{i_k}$  in  $T_m$  is marked with a 1 in  $\psi(A)$ . First note that  $\{[\psi(A)](\sigma^{i_k}) = 1\} \subset \Theta^{i_k}$ , because the algorithm fills all positions above a 1 with 0's. Next, fix  $a_{m+1}, a_{m+2}, \dots, a_{i_k-1} \in V_n$  such that

$$\{z^{\sigma^{m+1}} = a_{m+1}, \dots, z^{\sigma^{i_k-1}} = a_{i_k-1}\} \subset \Theta^{i_k} \cap \{[\psi(A)](\sigma^{i_1}) = \dots = [\psi(A)](\sigma^{i_k-1}) = 1\}$$

(we start at  $m+1$  because  $z^{\sigma^1}, \dots, z^{\sigma^m}$  are deterministic, equal to the points of  $A$ ). Then, conditioned on  $\{z^{\sigma^{m+1}} = a_{m+1}, \dots, z^{\sigma^{i_k-1}} = a_{i_k-1}\}$ , there are at least  $n-r$  possible positions for  $z^{\sigma^{i_k}}$ , and  $[\psi(A)](z^{\sigma^{i_k}}) = 1$  precisely when  $z^{\sigma^{i_k}} \in \mathcal{A}^{\sigma^{i_k}}$ , a set of size less than  $m+rm+\dots+r^g m$ .  $\blacksquare$

For  $i < g$  and  $B \subset T_m^i$ , write  $J(B) = \{\sigma' \in T_m : \sigma \rightarrow \sigma' \text{ for some } \sigma \in B\} \subset T_m^{i+1}$ . Again for  $i < g$ , a set of sets  $\mathcal{B} = (B_0, B_1, \dots, B_i)$  with  $B_0 \subset T_m^0$  and  $B_{j+1} \subset J(B_j) \forall j \in \{1, \dots, i-1\}$  will be called a *family* in  $T_m$ . Given  $\psi \in \{0, 1\}^{T_m}$ , a family  $\mathcal{B} = (B_0, \dots, B_i)$  in  $T_m$  is called  *$\psi$ -admissible* if

- (i.)  $\cup_{j=0}^i B_j \subset \{\psi = 0\}$ ;
- (ii.)  $|B_0| \geq \tilde{q}m$ ;  $\forall j \in \{1, \dots, i\}, |B_j| \geq \tilde{q} \cdot |J(B_{j-1}) \cap \{\psi = 0\}|$ .

$\mathcal{B}$  will be called  *$\psi$ -good* if it is  $\psi$ -admissible and

- (iii.)  $\forall j \in \{0, 1, \dots, i\}, |J(B_j) \cap \{\psi = 0\}| \geq (\tilde{q}r - 1 - \delta)(\tilde{q}r)^j m$ .

Finally, we say that  $\psi$  is *robust* if all  $\psi$ -admissible families are  $\psi$ -good. The next lemma shows the motivation for these definitions.

**Lemma 2.3** *There exist  $c, C > 0$  such that, for every  $n$  and  $A \subset V_n$ ,*

$$\mathbb{P}_n \left( |\hat{\xi}_g^A| < (1 + \delta)|A| \mid \psi(A) \text{ is robust} \right) \leq C e^{-c|A|}.$$

*Proof.* We define inductively a family  $\mathcal{B} = (B_0, \dots, B_{g-1})$ :

$$B_0 = \{\sigma \in T_{m,g}^0 : z^\sigma \text{ gives birth at time } 0\},$$

$$B_{j+1} = \{\sigma \in J(B_j) \cap \{\psi(A) = 0\} : z^\sigma \text{ gives birth at time } j+1\}.$$

The definition of  $B_0$  implies that  $\hat{\xi}_1^A \supset \{z^\sigma : \sigma \in J(B_0)\}$ . From the construction of  $\psi(A)$  we see that  $\sigma \mapsto z^\sigma$  is injective on  $J(B_0) \cap \{\psi(A) = 0\}$ , so we have  $|\hat{\xi}_1^A| \geq |J(B_0) \cap \{\psi(A) = 0\}|$ . Iterating this argument we get

$$|\hat{\xi}_j^A| \geq |J(B_{j-1}) \cap \{\psi(A) = 0\}|, \quad 1 \leq j \leq g. \quad (2.1)$$

From this equation with  $j = g$ , property (iii) in the definition of  $\psi(A)$ -good families and the choice of  $g$  in the beginning of this section, we see that the result will follow from

$$\mathbb{P}_n (\mathcal{B} \text{ is not } \psi(A)\text{-admissible} \mid \psi(A) \text{ is robust}) \leq C e^{-c|A|} \quad \forall n. \quad (2.2)$$

Define the events  $G_0 = \{|B_0| < \tilde{q}m\}$ ,  $G_j = \{|B_j| < \tilde{q} \cdot |J(B_{j-1}) \cap \{\psi(A) = 0\}|\}$ ,  $1 \leq j \leq g-1$ . We now have  $\{\psi(A) \text{ is robust}, \mathcal{B} \text{ is not } \psi(A)\text{-admissible}\} \subset \cup_{j=0}^{g-1} (\{\psi(A) \text{ is robust}\} \cap G_j)$ , so

$$\begin{aligned} \mathbb{P}_n (\mathcal{B} \text{ is not } \psi(A)\text{-admissible} \mid \psi(A) \text{ is robust}) &\leq \mathbb{P}_n(G_0 \mid \psi(A) \text{ is robust}) + \\ &\quad \sum_{j=1}^{g-1} \mathbb{P}_n \left( G_j \mid \psi(A) \text{ is robust}, \cap_{l=0}^{j-1} G_l^c \right). \end{aligned} \quad (2.3)$$

In order to bound the terms of the sum, we will need the estimate

$$x \in (0, 1) \implies \mathbb{P}(\text{Bin}(k, p) < xkp) \leq \exp\{-\gamma(x)kp\}, \quad (2.4)$$

where  $\gamma(x) = x \log x - x + 1$ . This follows from Markov's inequality; see Lemma 2.3.3 in [3]. We start noting that

$$\mathbb{P}_n(G_0 \mid \psi(A) \text{ is robust}) = \mathbb{P}(\text{Bin}(m, q) < \tilde{q}m) \leq \exp\{-\gamma(\tilde{q}/q)qm\}. \quad (2.5)$$

Next, on the event  $\{\psi(A) \text{ is robust}, \cap_{l=0}^{j-1} G_l^c\}$ , the family  $\mathcal{B}^j = (B_0, \dots, B_{j-1})$  is  $\psi(A)$ -admissible, hence  $\psi(A)$ -good because  $\psi(A)$  is robust, so  $|J(B_{j-1}) \cap \{\psi(A) = 0\}| > (\tilde{q}r - 1 - \delta)(\tilde{q}r)^{j-1}m > (\tilde{q}r - 1 - \delta)m$ . For  $G_j$  to occur, less than  $\tilde{q}|J(B_{j-1}) \cap \{\psi(A) = 0\}|$  can give birth at time  $j$ . Using (2.4), we get

$$\mathbb{P}_n\left(G_j \mid \psi(A) \text{ is robust}, \cap_{l=0}^{j-1} G_l^c\right) \leq \exp\{-\gamma(\tilde{q}/q)q(\tilde{q}r - 1 - \delta)m\}. \quad (2.6)$$

Putting (2.5) and (2.6) together back in (2.3), we get (2.2).  $\blacksquare$

**Lemma 2.4** *There exist  $\epsilon, d, D > 0$  such that*

$$m \leq \epsilon n \implies \mathbb{P}_n(\exists A \subset V_n : |A| = m, \psi(A) \text{ is not robust}) \leq De^{-dm} \quad \forall n.$$

*Proof.* Fix  $n \in \mathbb{N}$  and  $A \subset V_n$  with  $|A| = m$ . Let  $d_i = |\{\psi(A) = 1\} \cap T_m^i|$ ,  $1 \leq i \leq g$  and  $d = \sum d_i$ . Let  $i < g$  and  $\mathcal{B} = (B_0, \dots, B_i)$  be a  $\psi(A)$ -admissible family. We have

$$\begin{aligned} |B_0| &\geq \tilde{q}m; \\ |J(B_0) \cap \{\psi = 0\}| &\geq \tilde{q}rm - d_1 \geq \tilde{q}rm - d; \\ |B_1| &\geq \tilde{q}^2rm - \tilde{q}d_1; \\ |J(B_1) \cap \{\psi = 0\}| &\geq (\tilde{q}r)^2m - \tilde{q}rd_1 - d_2 \geq (\tilde{q}r)^2m - \tilde{q}rd; \\ &\dots \\ |J(B_i) \cap \{\psi = 0\}| &\geq (\tilde{q}r)^{i+1}m - (\tilde{q}r)^i d_1 - (\tilde{q}r)^{i-1}d_2 - \dots - \tilde{q}rd_{i-1} - d_i \\ &\geq (\tilde{q}r)^{i+1}m - (\tilde{q}r)^i d. \end{aligned}$$

Suppose that  $d \leq (1 + \delta)m$ . Then, for  $j \in \{0, \dots, i\}$  we have

$$|J(B_j) \cap \{\psi = 0\}| \geq (\tilde{q}r)^{j+1}m - (\tilde{q}r)^j d \geq (\tilde{q}r)^j(\tilde{q}r - 1 - \delta)m,$$

which means that  $\mathcal{B}$  is  $\psi(A)$ -good. We have thus shown that  $\{\psi(A) \text{ is not robust}\} \subset \{d > (1 + \delta)m\}$ . From this we get

$$\begin{aligned} \mathbb{P}_n(\exists A \subset V_n : |A| = m, \psi(A) \text{ is not robust}) &\leq \sum_{A: |A|=m} \mathbb{P}_n(\psi(A) \text{ is not robust}) \\ &\leq \sum_{A: |A|=m} \sum_{d=\lceil (1+\delta)m \rceil}^{(1+r+\dots+r^g)m} \sum_{D \subset T_m: |D|=d} \mathbb{P}_n([\psi(A)](\sigma) = 1 \forall \sigma \in D). \end{aligned}$$

We now bound  $|\{D \subset T_m : |D| = d\}|$  by  $2^{|T_m|}$  and use Lemma 2.2 to bound the probability; the above is then less than

$$\begin{aligned} & \binom{n}{m} (1+r+\dots+r^g) m 2^{(1+r+\dots+r^g)m} \left( \frac{(1+r+\dots+r^g)m}{n-r} \right)^{(1+\delta)m} \\ & \leq \left( \frac{ne}{m} \right)^m C^m \left( \frac{m}{n} \right)^{(1+\delta)m} \left( \frac{n}{n-r} \right)^{(1+\delta)m} \leq \left( C \left( \frac{m}{n} \right)^\delta \right)^m; \end{aligned}$$

here  $C$  is a constant that only depends on  $r, g$  and  $\delta$ , and whose value has changed in the last inequality. Now it suffices to choose  $\epsilon$  such that  $C\epsilon^\delta < 1$ .  $\blacksquare$

*Proof of Theorem 1.1.* From here on the proof continues very similarly to [1]; we present it for completeness. Fix  $\eta > 0$ ,

- take  $a, b$  corresponding to  $\eta$  as in Lemma 2.1;
- take  $c, C$  as in Lemma 2.3 and  $\epsilon, d, D$  as in Lemma 2.4;
- define  $F = C + D$ ,  $f = c \wedge d$ ,  $I = \lfloor e^{(f/2)^n} \rfloor$ ;
- assume that  $n$  is large enough so that  $n^b < \epsilon n < I$  and  $\delta n^b > 1$ .

Define

$$\begin{aligned} s_i &= \lceil a \log n \rceil + ig, \\ \alpha_i &= (\lfloor n^b \rfloor + i) \wedge \lfloor \epsilon n \rfloor, \quad 0 \leq i \leq I. \end{aligned}$$

Given  $A \subset V_n$  and  $\alpha \in \mathbb{N}$ , let  $\Pi_\alpha(A)$  denote the first  $\alpha$  elements of  $A$  (with respect to the order of  $V_n$ ). For  $x \in V_n$ , define

$$\zeta_0^x = \Pi_{\alpha_0}(\hat{\xi}_{s_0}^x), \quad \zeta_{i+1}^x = \Pi_{\alpha_{i+1}}(\hat{\xi}_{s_{i+1}}^{\zeta_i^x, s_i}), \quad 0 \leq i \leq I,$$

where  $\hat{\xi}_{s_{i+1}}^{\zeta_i^x, s_i}$  denotes the set of vertices that at time  $s_{i+1}$  descend from  $\zeta_i^x$  at time  $s_i$ . Finally, define the events

$$F_i^x = \{\psi(\zeta_i^x) \text{ is robust}\}, \quad G_i^x = \left\{ \left| \hat{\xi}_{s_{i+1}}^{\zeta_i^x, s_i} \right| \geq (1+\delta)|\zeta_i^x| \right\}, \quad 0 \leq i \leq I;$$

$$H_{-1}^x = \{\hat{\xi}_{s_0}^x > n^b\}, \quad H_i^x = H_{i-1}^x \cap F_i^x \cap G_i^x, \quad 0 \leq i \leq I;$$

$$H = \cap_{x \in V_n} ((H_{-1}^x)^c \cup H_I^x).$$

Fix  $i \geq 0$  and assume that  $H_i^x$  occurs. Then,  $|\zeta_0^x| = \alpha_0$  and, by the definition of  $G_0^x$ , we have  $|\hat{\xi}_{s_1}^{\zeta_0^x, s_0}| \geq (1+\delta)|\zeta_0^x| > |\zeta_0^x|$  because of the hypothesis  $\delta n^b > 1$ . So we have  $|\zeta_1^x| = \alpha_1$ , and arguing similarly we get  $|\zeta_j^x| = \alpha_j$  for  $1 \leq j \leq i+1$ . Since  $\zeta_j^x \subset \hat{\xi}_{s_j}^x$  for each  $j$ , we get  $|\hat{\xi}_{\lceil a \log n \rceil + jg}^x| \geq |\zeta_j^x| = \alpha_j$  for each  $j$ . As a consequence,

$$H_I^x \subset \left\{ \hat{\xi}_I^x \neq \emptyset \right\} \tag{2.7}$$

since  $\lceil a \log n \rceil + (I+1)g \geq I$ . We then have

$$\mathbb{P}_n \left( \left| \{x \in V_n : \hat{\xi}_I^x \neq \emptyset\} \right| > (\rho - \eta)n \right) \geq \mathbb{P}_n \left( H \cap \left\{ \left| \{x \in V_n : |\hat{\xi}_{s_0}^x| > \alpha_0\} \right| > (\rho - \eta)n \right\} \right),$$

the reason being that, on the event in the second probability, at least  $(\rho - \eta)n$  vertices reach time  $s_0$  with more than  $\alpha_0$  descendants, and all that do so continue having more than  $\alpha_i$  descendants at times  $s_i$ , for all  $i \leq I$ , and in particular are alive at time  $I$ , as seen in (2.7).

We know from Lemma 2.1 that, as  $n \rightarrow \infty$ ,

$$\mathbb{P}_n \left( \left| \{x \in V_n : |\hat{\xi}_{s_0}^x| > \alpha_0\} \right| > (\rho - \eta)n \right) \rightarrow 1.$$

Using Lemmas 2.3 and 2.4, we also have

$$\begin{aligned} \mathbb{P}_n(H^c) &\leq \sum_{x \in V_n} \sum_{i=0}^I [\mathbb{P}_n(H_{i-1}^x \cap (F_i^x)^c) + \mathbb{P}_n(H_{i-1}^x \cap (G_i^x)^c)] \\ &\leq \sum_{x \in V_n} \sum_{i=0}^I [\mathbb{P}_n(\exists A \subset V_n : |A| = \alpha_i, \psi(A) \text{ is not robust}) \\ &\quad + \mathbb{P}_n \left( \left| \hat{\xi}_{s_{i+1}}^{x, s_i} \right| < (1 + \delta) |\zeta_i^x| \mid \psi(\zeta_i^x) \text{ is robust} \right)] \\ &\leq \sum_{x \in V_n} \sum_{i=0}^I [Ce^{-c\alpha_i} + De^{-d\alpha_i}] \leq n \left( (\epsilon n - n^b) F e^{-f \lfloor n^b \rfloor} + (I - \epsilon n) F e^{-f \lfloor \epsilon n \rfloor} \right) \end{aligned}$$

which tends to 0 as  $n \rightarrow \infty$ . We thus have

$$\lim_{n \rightarrow \infty} \mathbb{P}_n \left( \left| \{x \in V_n : \hat{\xi}_I^x \neq \emptyset\} \right| > (\rho - \eta)n \right) = 1.$$

Using duality and attractiveness, we obtain

$$\inf_{0 \leq t \leq I} \mathbb{P}_n \left( \frac{|\xi_t^{V_n}|}{n} \geq \rho - \eta \right) = \mathbb{P}_n \left( \frac{|\xi_I^{V_n}|}{n} \geq \rho - \eta \right) = \mathbb{P}_n \left( \frac{|\{x : \hat{\xi}_I^x \neq \emptyset\}|}{n} \geq \rho - \eta \right) \rightarrow 1,$$

completing the proof.

## References

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